

CLAIMS

1. Method of keying, in a space presenting two spatial dimensions and one temporal dimension, a signal S measured in positions U subject to an uncertainty, from a set of N signals measured in determined positions, the N + 1 signals having their temporal origin in a same plane, the said method involving: - re-sampling the N + 1 signals in order to place them all in an identical sampling range,

- filtering the signal S in order to place it in a range of frequencies that is identical to that of the N signals,

and wherein the method also involves:

- defining for each position U associated with the measurements of the signal S a same neighbourhood of places V in the spatio-temporal space centred on the position U,

- producing a layered neural network RN_v for each location V in the neighbourhood of U, each network having an entry vector of dimension N associated with the measurements of the N signals and a scalar exit associated with a measurement of the signal S,

- for each neural network RN_v , defining a learning set such that the entries are the collection of all the vectors of measurements of the N signals situated at the locations V and the exits are the collection of the values of the signal S at the positions U for all the positions U,

- fixing a predetermined number of iterations N_{it} for all the neural networks and launching the learning phases of all the networks,

- for each neural network RN_v , calculating the value of the integral I_v of the function giving the error committed by the network at each iteration, from iteration 1 to iteration N_{it} .

- for each surface spatial position V_k of the neighbourhood with coordinates (X_k, Y_k, t_0) , selecting in the time dimension the pair of locations $V_{1k}(X_k, Y_k, t_1)$, $V_{2k}(X_k, Y_k, h)$, of the neighbourhood which correspond to the two smallest local minima of the two integrals (LV, LV_{2k}) ,

- for each surface spatial position V_k of the neighbourhood, retaining from among the two positions $V_h(X_k, Y_k, t_1)$, $V_{2k}(X_k, Y_k, h)$ the position V_m , for which the signal estimated by the respective neural networks RNV and RNV_{2k} presents a maximum variance,

- choosing from among the positions V_m the position V_{eal} for which the integral tV_m is minimum.

2. Method according to claim 1, wherein the use of the neural networks comprises:

- defining for each position U associated with the measurements of the signal S a same neighbourhood of places V in the spatio-temporal space centred on the position U ,

- producing a layered neural network RNV for each location V in the neighbourhood of U . each network having an entry vector of dimension $N \times M$ associated with the measurements on a time window of size M centred on V of the N signals and a scalar exit associated with a value of the signal S ,

- for each neural network, defining a learning set such that the entries are the collection of all the vectors of measurements taken in a time window of size M centred on V for the N signals and the exits are the collection of the values of the signal S at the positions U for all the positions U ,

- fixing a predetermined number of iterations N_{it} for all the neural networks and launching the learning phases of all the networks,

- for each neural network RN_v , calculating the value of the integral I_v of the function giving the error committed by the network at each iteration, from iteration 1 to iteration N_{it} ,

- for each surface spatial position V_k of the neighbourhood with coordinates (X_k, Y_k, t_0) , selecting in the time dimension the pair of locations $V_{1k}(X_k, Y_k, t_1)$, $V_{2k}(X_k, Y_k, t_2)$, of the neighbourhood which correspond to the two smallest local minima of the two integrals (I_{v1k}, I_{v2k}) .

- for each surface spatial position V_k of the neighbourhood, retaining from among the two positions $V_{1k}(X_k, Y_k, t_1)$, $V_{2k}(X_k, Y_k, t_2)$ the position V_m , for which the signal estimated by the respective neural networks RN_{v1} and RN_{v2k} presents a maximum variance,

- choosing from among the V_m positions the position V_{eal} for which the integral $I_{v_{eal}}$ is minimum.